# Machine Learning & Data Mining Lecture 4

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### Outline

- Decision Tree Regression Model
- Random Forest Regression Model



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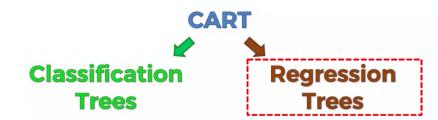
# **Decision Tree Regressions**

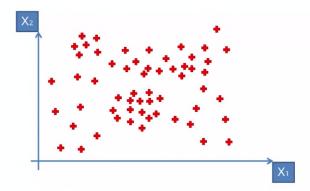
### CART

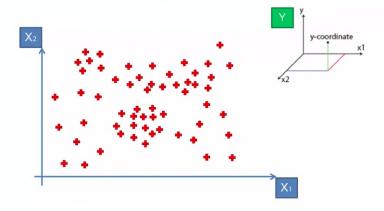


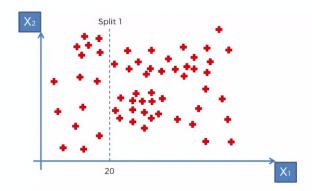


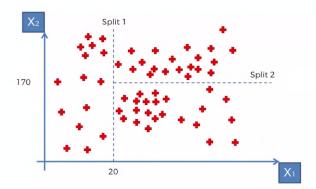
Classification Trees Regression Trees

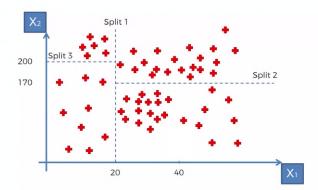


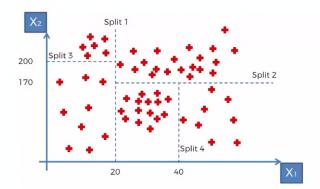




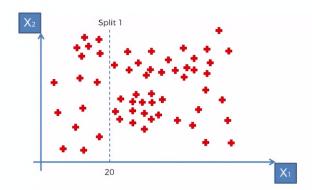




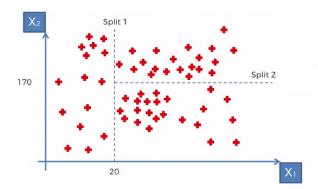


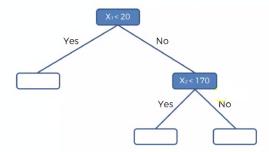


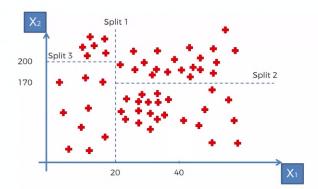
# Rewind...

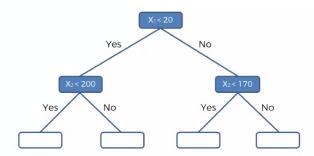


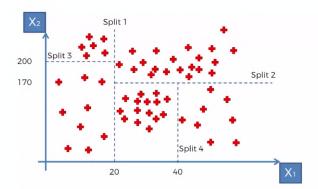


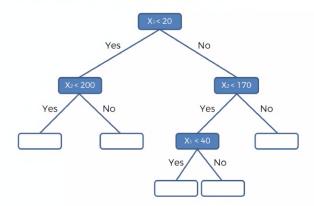


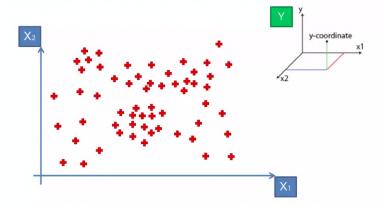


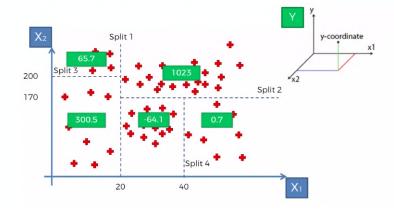


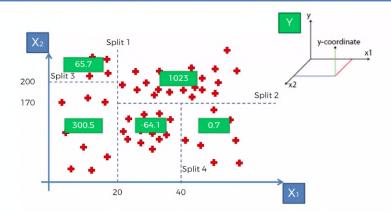




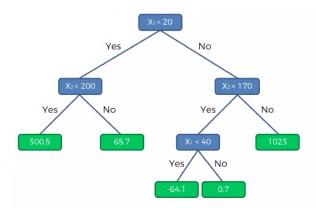








Example :  $x_1 = 50 \ x_2 = 10$ 



### Making the Splits

#### How are the splits done?

https://www.youtube.com/watch?v=g9c66TUyIZ4

# **Random Forest Regression**

# **Ensemble Learning**

STEP 1: Pick at random K data points from the Training set.

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STEP 2: Build the Decision Tree associated to these K data points.

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STEP 3: Choose the number Ntree of trees you want to build and repeat STEPS 1 & 2.

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STEP 4: For a new data point, make each one of your Ntree trees predict the value of Y to for the data point in question, and assign the new data point the average across all of the predicted Y values.

# **Feature Scaling**

Standardisation	Normalisation
$x_{\text{stand}} = \frac{x - \text{mean}(x)}{\text{standard deviation }(x)}$	$x_{\text{norm}} = \frac{x - \min(x)}{\max(x) - \min(x)}$

$$\mu = 0$$
  $\sigma = 1$ 

When to use one versus the other?

#### **Standardization**

- When the distribution of data is unknown
- When model is sensitive to different scales (SVR)

- When the distribution of features is close to norma
- When model assumes normal distribution (e.g. Linear Regression Logistic Regression, etc.)

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### Let's get Started!

Access Google Colaboratory through your Gmail account